

Lecture20

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Lecture20

Linear Algebra

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(Department of CE)

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Chapter5

Eigenvalues and Eigenvectors

(Department of CE)

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Review: Eigenvectors and eigenvalues

Definition

An eigenvector or characteristic vector v for a square matrix A is a nonzero vector that changes at most by a scalar factor when that matrix is applied to it, i.e., $Av = \lambda v$. The corresponding eigenvalue λ is denoted by as an eigenvalue of A .

Review: solving $Ax = \lambda x$

\wedge

$$\Leftrightarrow (\lambda I - A)x = 0 \quad \Leftrightarrow \lambda \in \text{eigenvalues of } A$$

$x \neq 0$

$$\det(\lambda I - A) = 0$$

- Find the roots of the polynomial $\det(A - \lambda I) = 0$. These roots are the eigenvalues of A .
- The sum of the n eigenvalues equals the sum of the n diagonal entries:

$$\text{Trace of } A := a_{11} + \cdots + a_{nn} = \lambda_1 + \cdots + \lambda_n.$$

- The product of the n eigenvalues equals the determinant of A , that is

$$\lambda_1 \times \cdots \times \lambda_n = \det A.$$

$$\underline{\underline{P(\lambda) = \det(\lambda I - A)}} \\ \lambda_1$$

Example

- Find the eigenvalues and eigenvectors for diagonal matrices.

$$A = \begin{bmatrix} d_1 & & \\ & d_2 & \\ & & \ddots \\ & & & d_n \end{bmatrix}$$

$$E = \{e_1, \dots, e_n\}$$

Diagonalization of a Matrix

$$p(\lambda) = \det(\lambda I - A) = 0 \Rightarrow \det \begin{bmatrix} \lambda - 1/2 & 1/2 \\ 1/2 & \lambda - 1/2 \end{bmatrix}$$

- Example.** The eigenvector matrix of the projection $P = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}$

$$\text{is } S = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}, \text{ and we have}$$

$$Pv_1 = 0 \quad \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 0 \quad S^{-1}PS = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

- The eigenvector matrix S converts A into its eigenvalue matrix which is diagonal.

$$Pv_2 = v_2 \quad \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad PS = S \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} \Rightarrow \lambda_1 = 0, \dots, \lambda_n = a_{nn}$$

90° rotation

- The eigenvalues themselves are not so clear for a rotation:

$$K = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$

$$\text{has } \det(K - \lambda I) = \lambda^2 + 1.$$

- The eigenvalues of K are imaginary numbers $\lambda = \pm i$, with eigenvectors $\begin{bmatrix} 1 \\ -i \end{bmatrix}$ and $\begin{bmatrix} 1 \\ i \end{bmatrix}$.
- The eigenvectors are also not real.
- The eigenvalues are distinct, even if imaginary, and the eigenvectors are independent. They go into the columns of S :

$$\begin{bmatrix} 1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} -i & 0 \end{bmatrix}$$

$$\dim V = n$$

$$T: V \rightarrow V$$

$$A = [T]_B$$

$$T(x) = \lambda x$$

$$B = \{v_1, \dots, v_n\}$$

$$[T]_B = \begin{bmatrix} a_{11} & & \\ & \ddots & \\ & & a_{nn} \end{bmatrix}$$

$$A = \begin{bmatrix} a_{11} & & * \\ & \ddots & \\ 0 & & a_{nn} \end{bmatrix}$$

$$\det(\lambda I - A) = 0$$

$$\det \begin{bmatrix} \lambda - a_{11} & & * \\ & \ddots & \\ 0 & & \lambda - a_{nn} \end{bmatrix}$$

$$= (\lambda - a_{11}) \dots (\lambda - a_{nn})$$

$$\Rightarrow \lambda_1 = a_{11}, \dots, \lambda_n = a_{nn}$$

$$P^2 = P \quad P \in M_n(\mathbb{R})$$

$$S^{-1}AS = B$$

$$p(\lambda) = \lambda^{\alpha} (\lambda - 1)^{\beta}$$

$$\alpha + \beta = n$$

$$\text{Rank } P = r$$

$$p(\lambda) = \lambda^{n-r} (\lambda - 1)^r$$

$$N(P) = \text{Span} \{v_1, \dots, v_{n-r}\}$$

$$\dim N(P) = n - r$$

$$N(P)$$

- The eigenvalues are distinct, even if imaginary, and the eigenvectors are independent. They go into the columns of S :

$$S = \begin{bmatrix} 1 & 1 \\ -i & i \end{bmatrix} \quad S^{-1}KS = \begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix} \quad B$$

Linear Spaces on \mathbb{C}

A Linear space is a set V along with an **addition** on V and a **scalar multiplication** on \mathbb{C}

$$(V, +, \cdot)$$

such that the following properties hold:

- commutativity** $u + v = v + u$ for all $u, v \in V$;
- associativity** $(u + v) + w = u + (v + w)$ and $(ab)v = a(bv)$ for all $u, v, w \in V$ and all $a, b \in \mathbb{C}$;
- additive identity** there exists an element $0 \in V$ such that $v + 0 = v$ for all $v \in V$;
- additive inverse** for every $v \in V$, there exists $w \in V$ such that $v + w = 0$;
- multiplicative identity** $1v = v$ for all $v \in V$;
- distributive properties** $a(u + v) = au + av$ and $(a + b)u = au + bu$ for all $a, b \in \mathbb{C}$ and all $u, v \in V$.

Diagonalizable matrices

Definition

Similarity. Let $A, B \in M_n(\mathbb{F})$, $\mathbb{F} = \mathbb{R}$ or \mathbb{C} . We say that A and B are **similar** over \mathbb{F} , if there exists an invertible matrix $S \in \mathbb{F}$ such that $S^{-1}AS = B$.

Definition

Assume $A \in M_n(\mathbb{R})$. A is called **diagonalizable** if it is similar to a diagonal matrix, i.e., if there exists an invertible matrix S and a diagonal matrix D such that

$$S^{-1}AS = D$$

Diagonalization of a Matrix

Theorem

The Diagonalization Theorem. Let $A \in M_n(\mathbb{F})$ where $\mathbb{F} = \mathbb{R}$ or \mathbb{C} . The matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

$$= \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$

$$N(p)$$

$$P \in M_n(\mathbb{R})$$

$$B = \{v_1, \dots, v_{n-r}, v_{n-r+1}, \dots, v_n\}$$

$N(p)$

$\mathbb{R}^n \supset \mathbb{R}^{n-r} \supset \mathbb{R}^r$

$$T: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

$$T(x) = P_{\mathbb{R}^r}$$

$$[T]_B = \begin{bmatrix} 0 & C \\ 0 & D \end{bmatrix}$$

$$B \text{ is a basis for } \mathbb{R}^n \Rightarrow B^{-1}TB = \begin{bmatrix} 0 & C \\ 0 & D \end{bmatrix}$$

$$T(v) = 0 \Rightarrow T(v) = 0$$

$$\det(\lambda I - P) = \det(\lambda I - [T]_B)$$

$$\det(\lambda I - P) = \det(\lambda I - [T]_B)$$

$$\det(\lambda I - D) = 0$$

$$Dx = 0$$

$$[T]_B \begin{bmatrix} 0 \\ x \end{bmatrix} = \begin{bmatrix} Cx \\ Dx \end{bmatrix} = 0 \Rightarrow Dx = 0$$

$$\Rightarrow x = 0 \Rightarrow \text{مصفوفة صفرية}$$

$$N(p) = \{v_1, \dots, v_{n-r}\}$$

$$P(\lambda) = \lambda^{n-r} (\lambda - 1)^r$$

$$n-r \leq \dim N(p) \leq \frac{n}{p} \Rightarrow \lambda = 1$$

$$A \quad \lambda_1, \dots, \lambda_k$$

$$N_k = \{x \mid P(x) = \lambda_k x\}$$

$(\lambda - \lambda_k)$

The matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

$$S^{-1}AS = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$

$$AS = S \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} \Rightarrow A[v_1 \dots v_n] = [v_1 \dots v_n] \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$

(\Leftrightarrow) زن کند A خود را به سبب S که به سبب S می باشد

$$S = [v_1 \dots v_n]$$

$$\Rightarrow v_i: A v_i = \lambda_i v_i$$

$$[\lambda_1 v_1 \dots \lambda_n v_n]$$

v_i به سبب v_i

زیر $v_i \neq 0$ (S درستی)

λ_i به سبب λ_i

$\{v_1, \dots, v_n\}$ به سبب v_i

$$d_i v_i = \sum_{j=1}^n c_j \lambda_j v_j$$

$$A v_i = \sum_{j=1}^n c_j A v_j$$

$$= \sum_{j=1}^n c_j \lambda_j v_j$$

$$\Rightarrow \sum_{j=1}^n c_j (\lambda_i - \lambda_j) v_j = 0$$

همان v_1, \dots, v_n به سبب v_i

$$c_i (\lambda_i - \lambda_{i+1}) = 0$$

$$\text{و } c_i \neq 0 \text{ و } v_i \neq 0$$

$$c_i (\lambda_i - \lambda_{i+1}) = 0$$

\downarrow

$$\lambda_i - \lambda_{i+1} = 0$$

\downarrow

$$\forall i < n \quad \lambda_i = \lambda_{i+1} \quad \times$$

A corollary of Diagonalization theorem

Fact

If the eigenvalues of A are distinct from each other, then A is diagonalizable.

$$A = \lambda = \lambda_1 = \lambda_2 = \dots = \lambda_n$$

همان $\lambda_1, \dots, \lambda_n$ به سبب λ_i

v_1, \dots, v_n به سبب v_i

همان v_1, \dots, v_n به سبب v_i

$$v_i = \sum_{j=1}^n c_j v_j \quad \text{و } c_i \neq 0$$

Remarks

$$f(\lambda) = \det \begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix} = \lambda^2 \Rightarrow \text{تقریباً در آن صفر}$$

- Not all matrices are diagonalizable.

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$Ax = 0$$

$$\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

- The diagonalizing matrix S is not unique.

$$Ax = 0 \Leftrightarrow x_2 = 0$$

$$W = \{x \mid Ax = 0\} = \left\{ \begin{bmatrix} x_1 \\ 0 \end{bmatrix} \mid x_1 \in \mathbb{R} \right\} \Rightarrow \dim W = 1$$

Diagonalizable linear transformation

Definition

A linear transformation $T : V \rightarrow V$ where the dimension of V is finite, is said to be diagonalizable if there exists a basis B such that $[T]_B$ is diagonal matrix.

$$[T]_B = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$

$$[T]_B = P^{-1} [T]_E P$$

Diagonalizable linear transformation

Theorem

Let $T : V \rightarrow V$ be a linear transformation where the dimension of V is finite with different eigenvalues $\lambda_1, \dots, \lambda_k$. Suppose that W_i is null space of $T - \lambda_i I$ for each $1 \leq i \leq k$. The the following statements are equivalent:

- ① T is diagonalizable.
- ② Its eigenvalue vector is $f(x) = (x - d_1)^{n_1} \dots (x - d_k)^{n_k}$ and $\dim W_i = n_i$.
- ③ $\sum_{i=1}^k \dim W_i = \dim V$.

$$\sum \dim W_i = \dim V$$

$$W_i = \{x \mid Ax = d_i x\}$$

قبر برای d_i

فضای d_i

Thank You!